



SMART Initiative

Science-informed Machine Learning to Accelerate
Real Time (SMART) Decisions in Subsurface Applications



U.S. DEPARTMENT OF
ENERGY



SMART Initiative



Primary Goals



Real-Time Visualization
"CT" for the Subsurface



Rapid Prediction
Virtual Learning



Real-Time Forecasting
"Advanced Control Room"

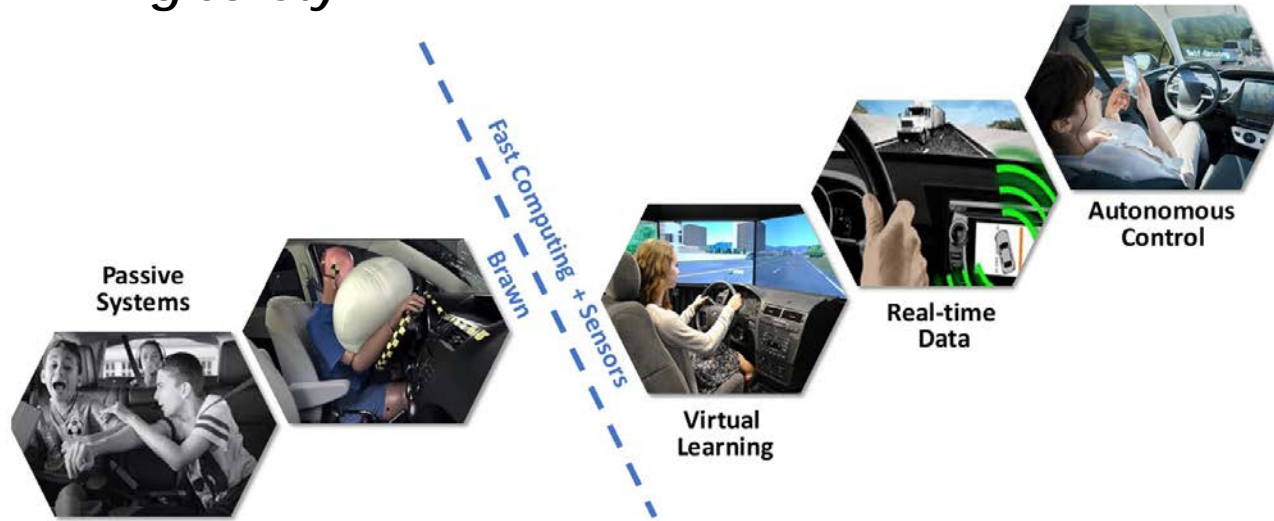
Technical Team



Science-informed Machine Learning to Accelerate Real Time (SMART) Decisions in Subsurface Applications

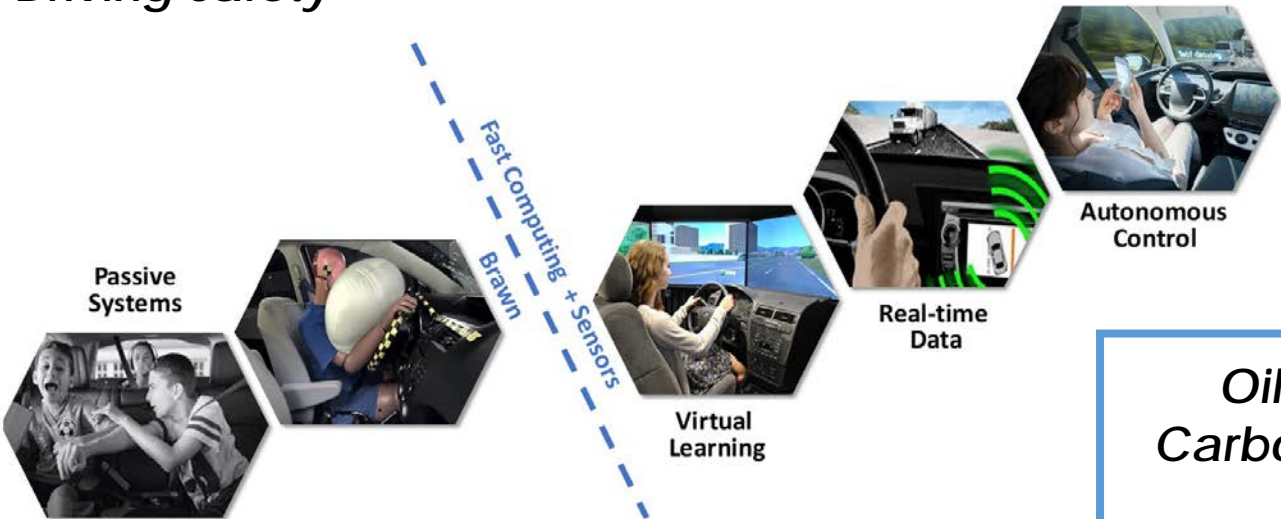
Transformational Experience

Driving Safety



Transformational Experience

Driving Safety



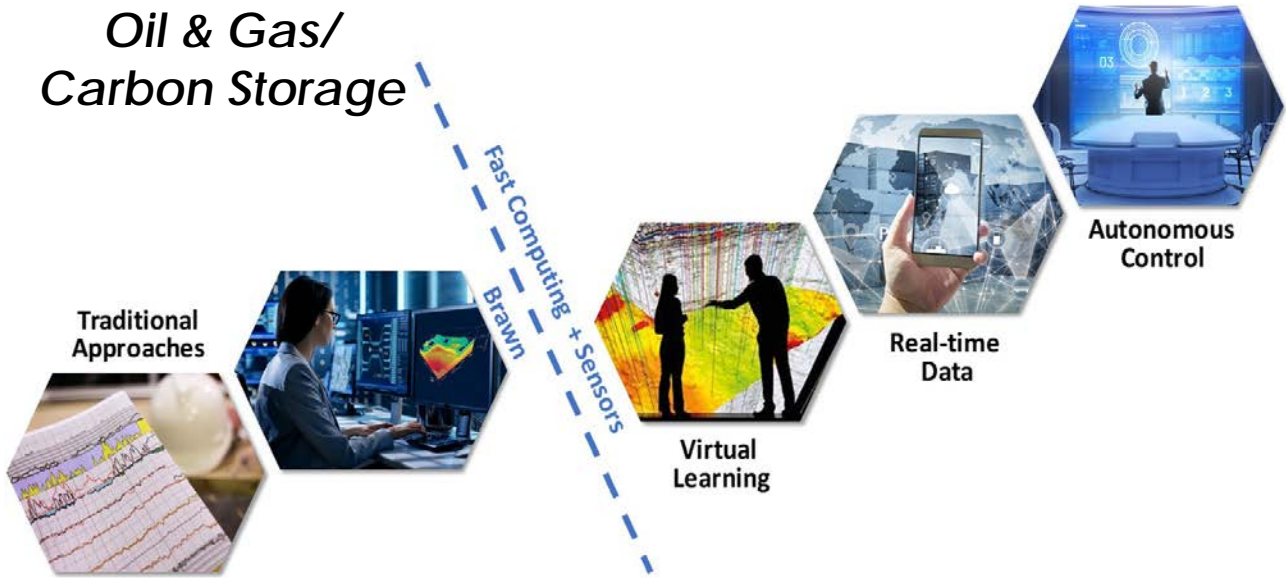
Pattern Recognition

Autonomous Monitoring

Virtual Reality Learning

Autonomous Control

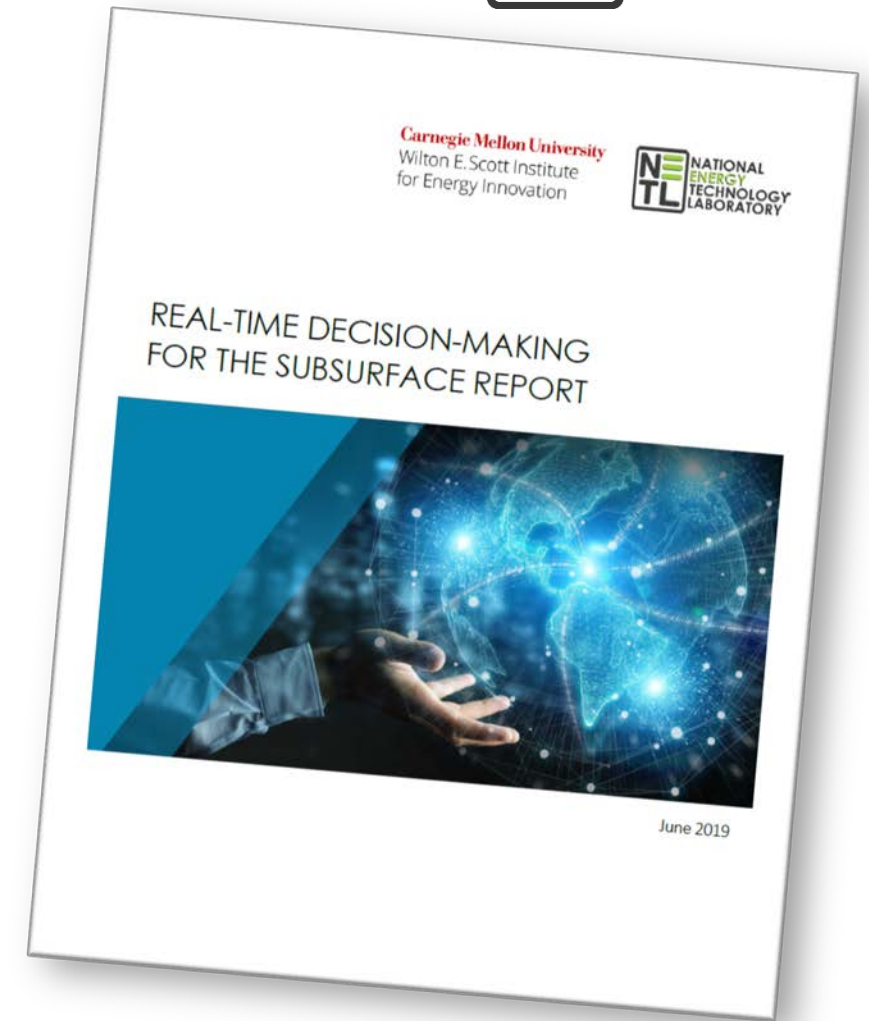
Oil & Gas/ Carbon Storage



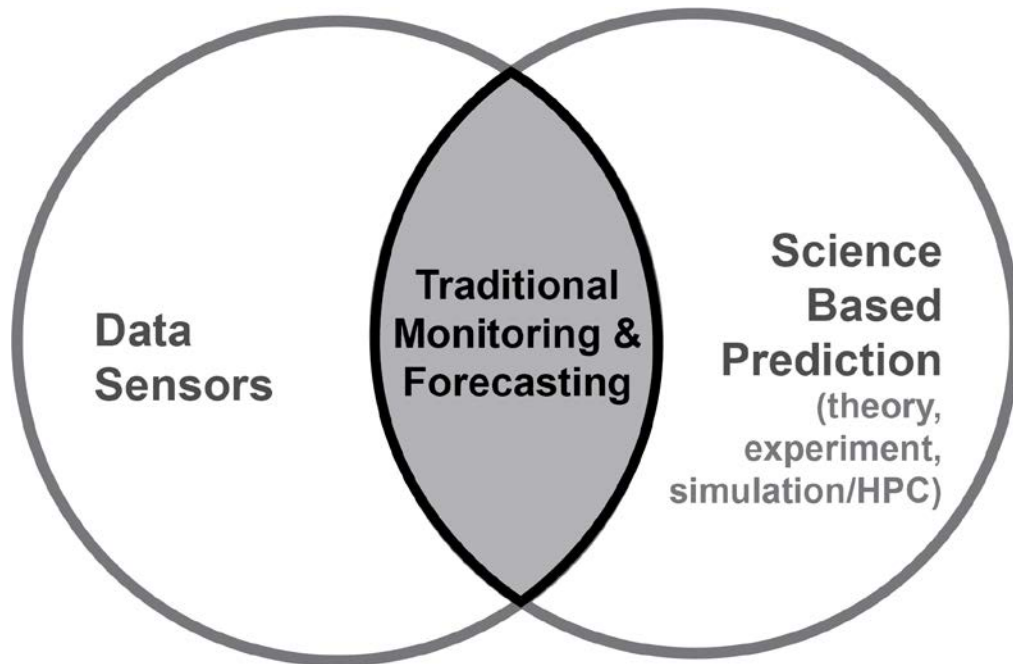
- Science-Informed
- Machine Learning
- Accelerating
- Real
- Time Decisions for Subsurface Applications

Programmatic Must-Haves...

1. Transformational Products & Application(s)
2. Quantifiable Goals
3. Early Wins
4. Industry Buy-in
5. Leverage Existing Capabilities

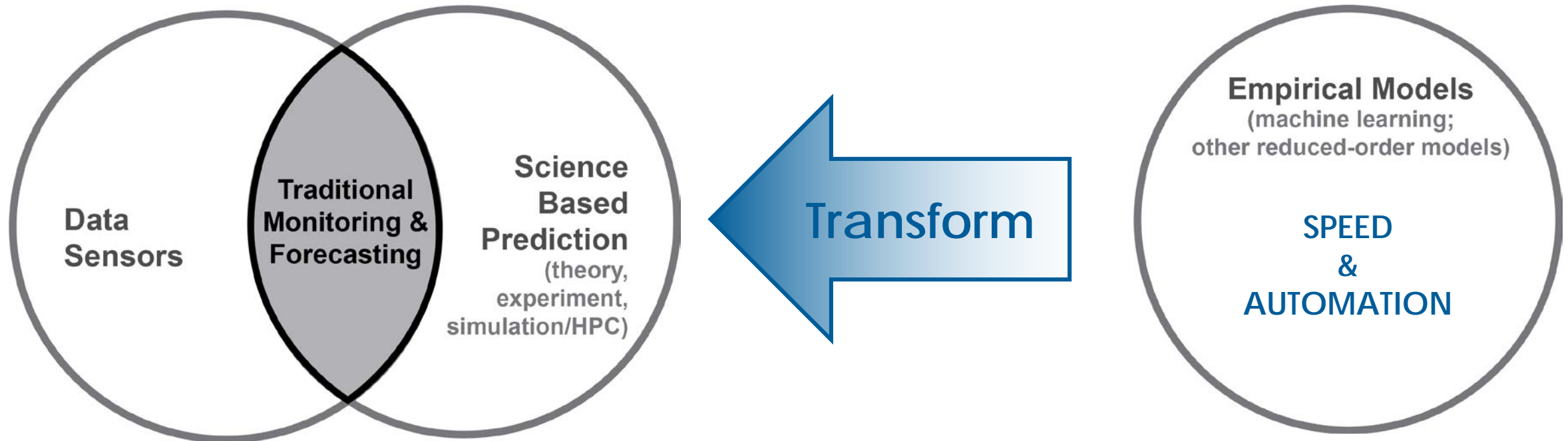


How can machine learning transform subsurface operations?

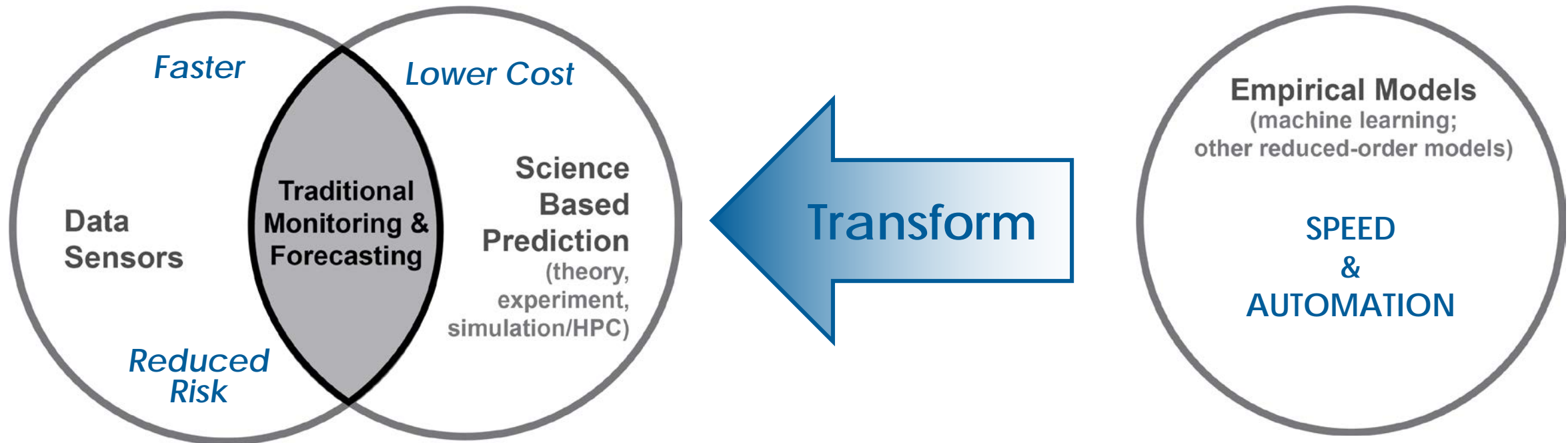


Traditional approach

How can machine learning transform subsurface operations?



How can machine learning transform subsurface operations?



Confluence of Data, Computational Capability, and Machine Learning



Real-Time Visualization
"CT" for the Subsurface



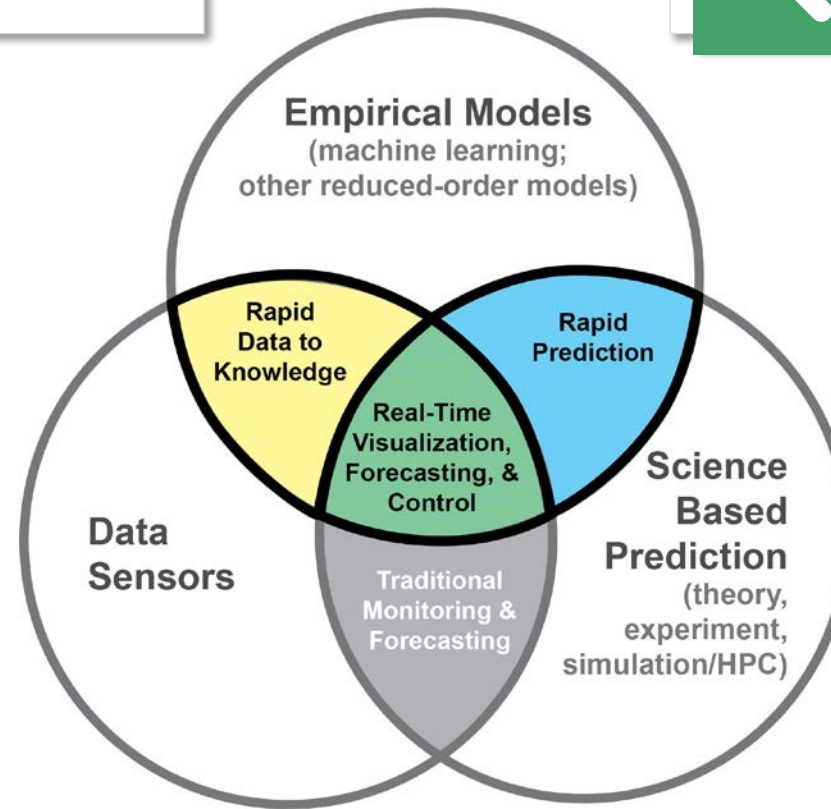
Real-Time Forecasting
"Advanced Control Room"



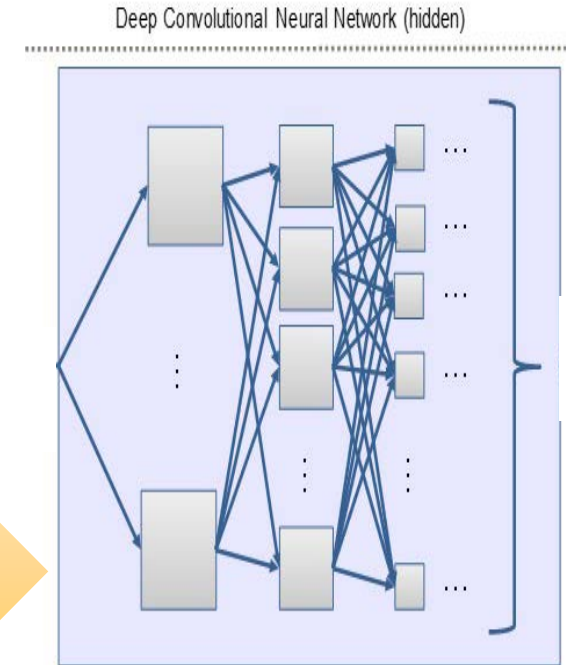
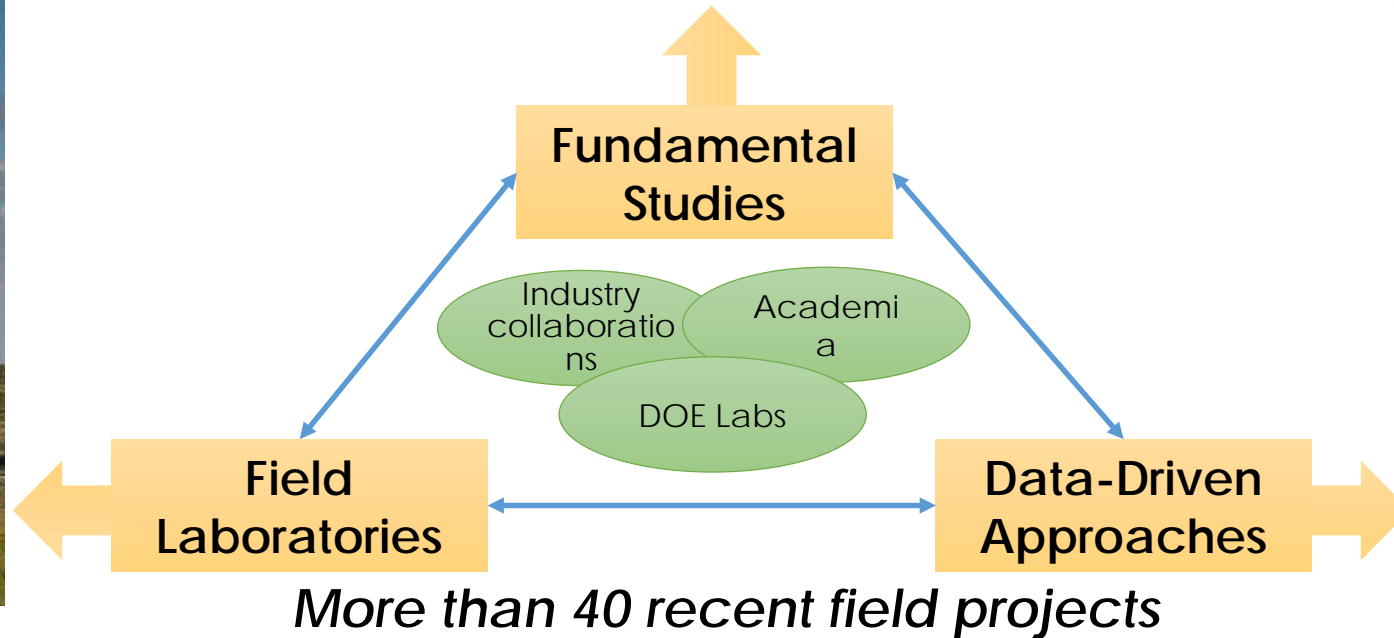
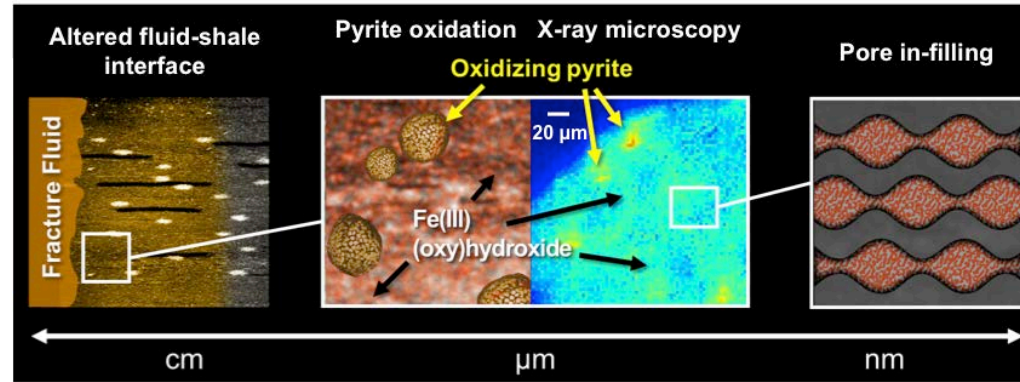
Rapid Data to Knowledge
Autonomous Monitoring



Rapid Prediction
Virtual Learning



Current and Prior Fossil Energy Investments Enabling Machine Learning for the Subsurface

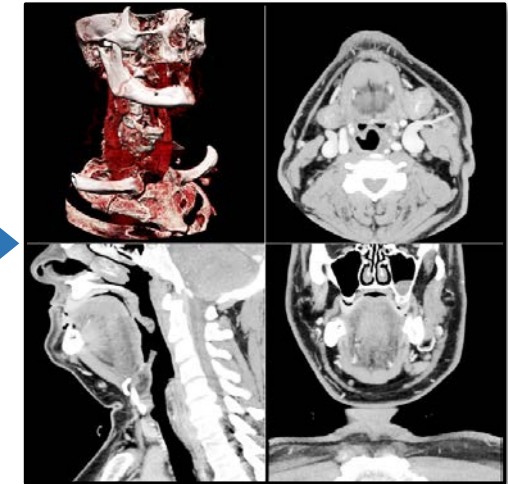
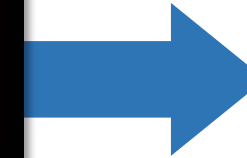
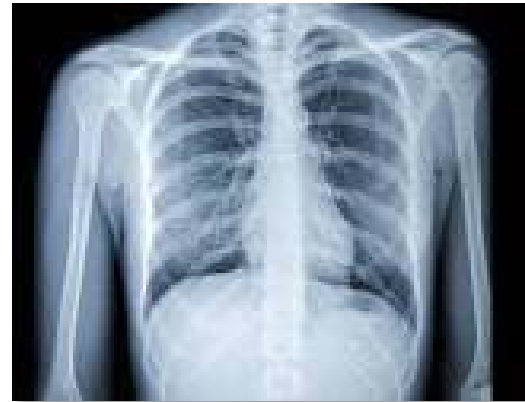


Real-Time Visualization



Real-Time Visualization *"CT" for the Subsurface*

Vision: Transform reservoir management via dramatic improvements in subsurface visualization, exploiting ML to achieve speed and enhanced detail.



Three Imaging Targets

1. Rock and fluid Properties
2. Pressure/stress
3. Faults and fracture networks

Potential Technology Pathways

- Joint inversion
- Multi-INT
- Tonal-Noise Tomography

Real-Time Forecasting

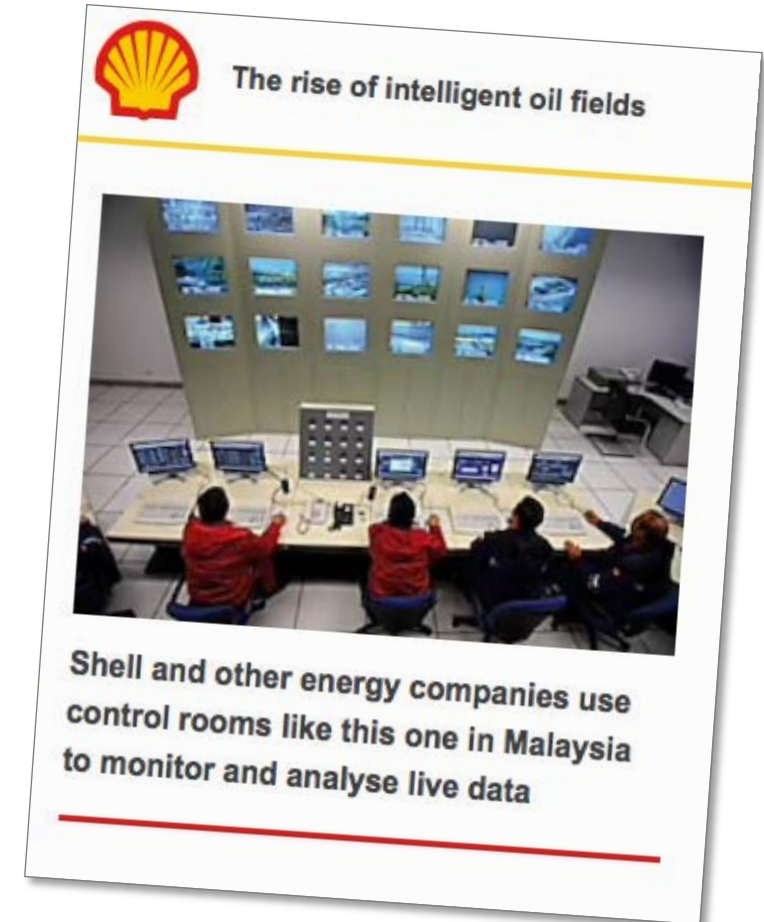


Real-Time Forecasting *"Advanced Control Room"*

Vision: Transform "human-in-the-loop" decisions on reservoir management by rapid visualization of forecasted behavior for different operational decisions.

Potential Operational Decisions

- How to adjust production rates and volumes in multiple wells to maximize recovery, sweep efficiency, ...
- **How to adjust CO₂ injection & brine production in multiple wells to maximize storage and minimize pressure plume**
- When and how to refrack an unconventional to increase total recovery



Shell.com



Rapid Prediction *Virtual Learning*

Vision: Enable a virtual learning environment for exploring and testing strategies to optimize reservoir development, management, & monitoring prior to field activities.

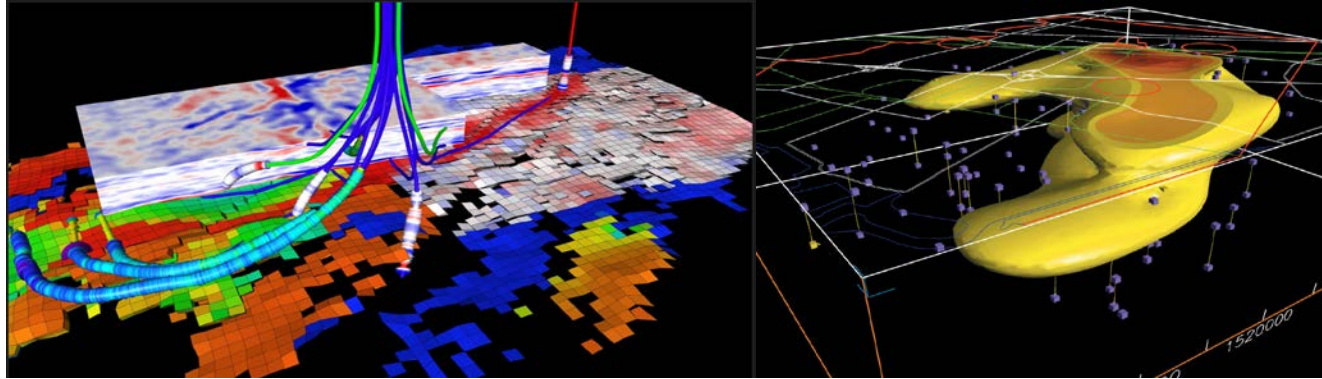


Virtual learning means experiential learning in a computer based environment that responds to a user's actions in real time, simulating the behavior of the subsurface system based on physics-based knowledge.

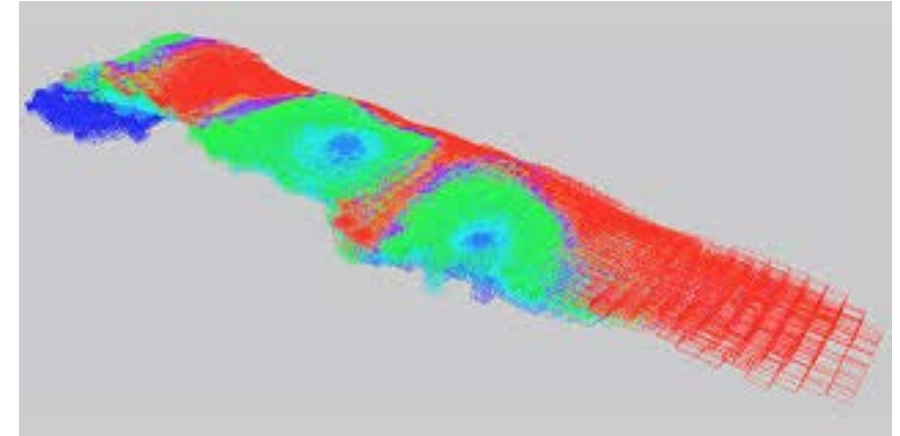
Physics-based knowledge means that the relevant subsurface processes must be known, well characterized, and able to be simulated with high fidelity.

Real-time is enabled by (1) coupling the high fidelity simulations with rapid, empirical methods (e.g., machine learning) and (2) exploiting developments for rapid visualization gaming environments.

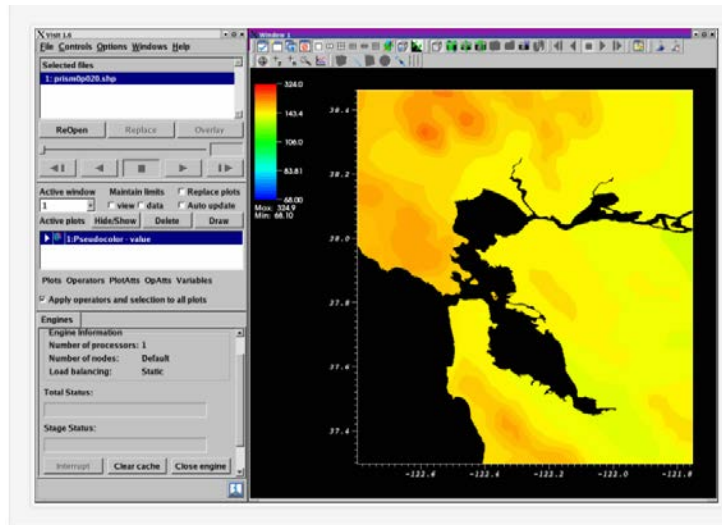
Visualization Capabilities



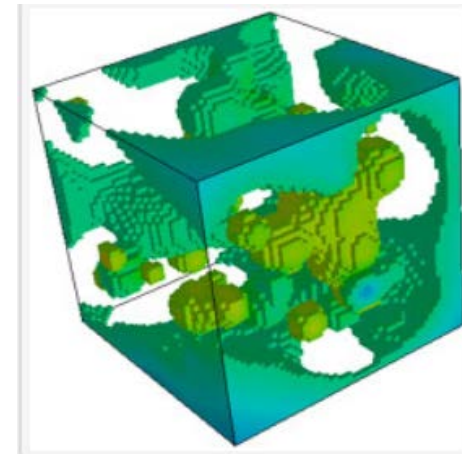
Dynamic Graphics



GEM



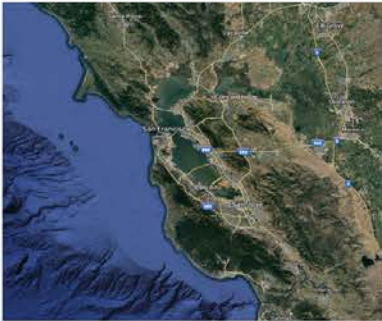
Visit



Analogy for new visualization functionality



Scale too coarse for planning route to stadium, but useful for setting context of where in the world you are going to be



Target scale : maps useful for planning route to stadium

Scale too fine for planning route, but useful for determining if and why you would want to go

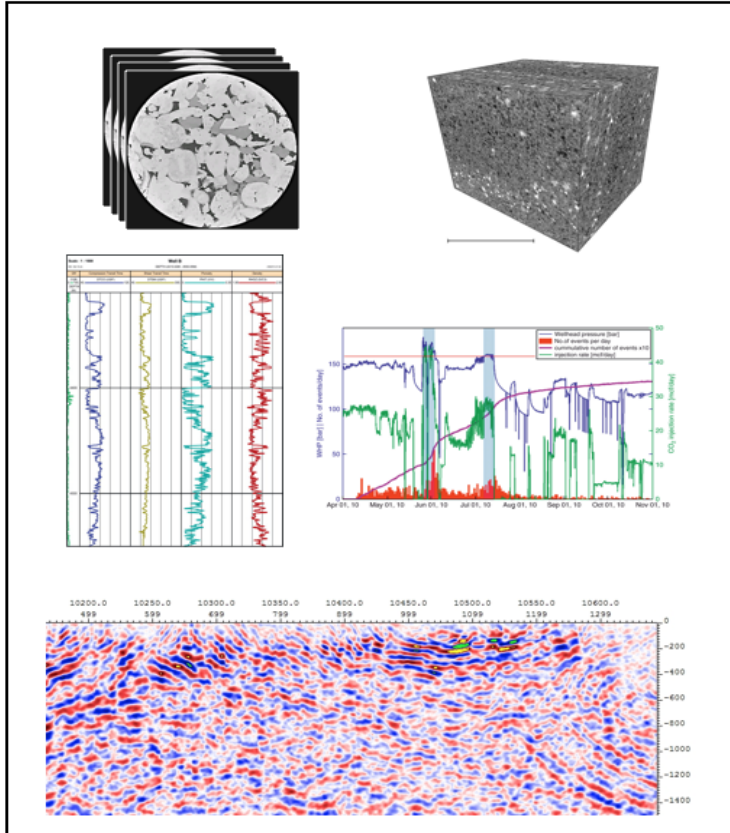


Decreasing Scale



Conceptual approach

Input



Physics-Based Processing

$$(Rf)(\theta, s) = \int_{x=\theta=s} f(x) dx$$

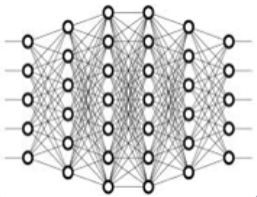
$$\Phi = \Phi_0 e^{-\pi i x}$$

$$\frac{\partial}{\partial t}(\rho_N \alpha_N) + \frac{\partial(\rho_N j_{Ni})}{\partial x_i} = I_N$$

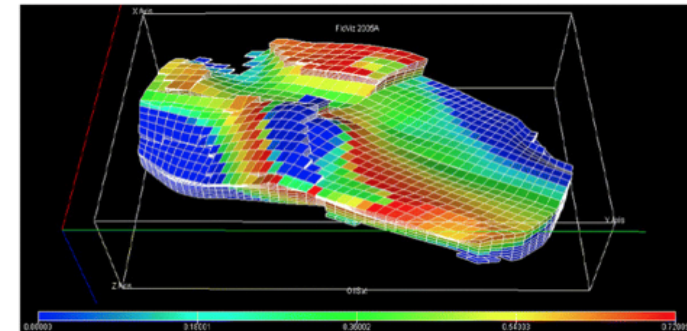
$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2} \quad f = -\nabla U^b(x)$$

Integrated With

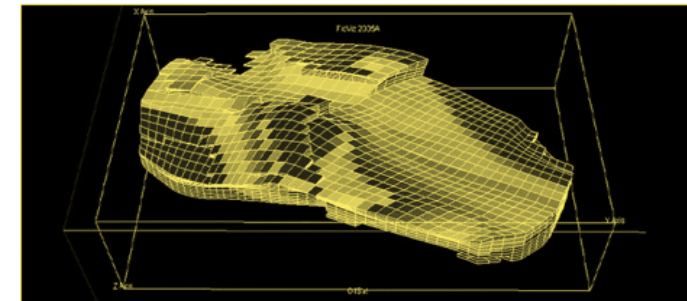
Machine Learning



Output



CO2 Saturation



Uncertainty map

Image credits: Miral Tawfik, Penn State University; Zuleima Karpyn, Penn State University; Raoof Gholami, Curtin University; Shawn Maxwell, IMA GE: Itasca Microseismic and Geomechanical Evaluation; Fathom Geophysics; Shattersock; Gholamreza Khademi, Cleveland State University

Goals for new visualization capabilities

- **Incorporates data and prediction**
 - Fast and accurate
- Both static and dynamic features
- Visualizes uncertainty
- Incorporates different scales
- Intuitive for the non-technical person

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What functionality do you need?

Thank You!

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